

# Noncommutative realizations: automatic analyticity and more

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# The functional calculus

Let  $f : (a, b) \rightarrow \mathbb{R}$ .

Given a self-adjoint matrix  $A$  with spectrum in  $(a, b)$  diagonalized by a unitary matrix  $U$ , that is,

$$A = U^* \begin{pmatrix} \lambda_1 & 0 & \dots \\ 0 & \lambda_2 & \dots \\ \vdots & \vdots & \ddots \end{pmatrix} U$$

we define the expression  $f(A)$  via the following formula.

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# Matrix inequalities

Let  $A$  and  $B$  be self-adjoint matrices:

- We say  $A \leq B$  if  $B - A$  is positive semi-definite.
- We say  $A < B$  if  $B - A$  is positive definite.

A positive definite matrix is a self-adjoint matrix with positive eigenvalues, and a positive semi-definite matrix is a self-adjoint matrix with nonnegative eigenvalues.

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# Special matrix functions

Let  $f : (a, b) \rightarrow \mathbb{R}$ .

- $f$  is **matrix monotone** if, for any natural number  $n \in \mathbb{N}$ , and any pair of  $n$  by  $n$  self-adjoint matrices  $A$  and  $B$  with spectrum in  $(a, b)$ ,

$$A \leq B \Rightarrow f(A) \leq f(B).$$

- $f$  is **matrix convex** whenever  $f$  evaluated on  $n$  by  $n$  matrices via the matrix functional calculus is a matrix-valued convex function. That is,

$$f\left(\frac{A+B}{2}\right) \leq \frac{f(A) + f(B)}{2}$$

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# Classical representations: Löwner-Nevanlinna

Let  $\Pi$  denote the complex upper half plane. A **Pick function** is an analytic map  $f : \Pi \rightarrow \bar{\Pi}$ .

## Theorem (Löwner's theorem)

*A function  $f : (a, b) \rightarrow \mathbb{R}$  is matrix monotone if and only if  $f$  analytically continues to a Pick function  $f : \Pi \cup (a, b) \rightarrow \bar{\Pi}$ .*

## Theorem (Nevanlinna's representation)

*$f$  is a Pick function if and only if there exist  $a \in \mathbb{R}$ ,  $b \geq 0$ , and a positive finite Borel measure  $\mu$  with  $\int \frac{1}{t^2+1} d\mu < \infty$  such that for all  $z \in \Pi$*

$$f(z) = a + bz + \int_{\mathbb{R}} \frac{1}{t-z} - \frac{t}{t^2+1} d\mu(t).$$

*Moreover,  $a, b, \mu$  are uniquely determined by  $f$ .*

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## Theorem (Kraus 1937)

Let  $f : (-1, 1) \rightarrow \mathbb{R}$ .  $f$  is matrix convex if and only if

$$f(x) = a + bx + \int_{[-1,1]} \frac{x^2}{1+tx} d\mu(t)$$

where  $a, b \in \mathbb{R}$  and  $\mu$  is a finite measure supported on  $[-1, 1]$ . Note that all such functions analytically continue to the upper half plane.

# A circle of ideas

- $f$  is matrix monotone on  $(a, b)$
- $\Rightarrow f$  can be approximated uniformly with smooth functions that preserve properties of  $f$  (mollifiers)
- $\Rightarrow f$  is  $C^1$
- $\Rightarrow f$  has an infinitely differentiable integral representation against a probability measure  $\nu$
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Löwner and Kraus theorems are example of what we might call a **automatic real analyticity theorem** –

matrix monotone  $\Rightarrow$  real analytic representation  
 $\Rightarrow$  analytic continuation to Pick function

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Q: Does this heuristic generalize?

Lots of realization/representation analogues to classical theorems in the nc setting. Samples we drew on

- Nevanlinna representations (Agler-T.D.-Young '16, Pascoe-T.D. '17)
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# The matrix universe

We define the **matrix universe** to be:

$$\mathcal{M}^d = \bigcup M_n(\mathbb{C})^d,$$

where  $M_n(\mathbb{C})^d$  is  $d$ -tuples of  $n$  by  $n$  matrices.

We define the **real matrix universe** to be:

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# Noncommutative sets

**Free set**  $D \subseteq \mathcal{M}$ :

- 1  $X, Y \in D \Rightarrow X \oplus Y \in D$
- 2  $X$  implies  $U^*XU \in D$  whenever  $U$  is unitary.

Examples: tuples of contractions, tuples of commuting contractions, block 2 by 2 self-adjoint contractions, etc.

We say  $D$  is **(open, connected, convex)** whenever each  $D \cap M_n(\mathbb{C})$  is (open, connected, convex.)

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# Butterfly realization - a Kraus theorem

## Theorem (Helton, McCullough, Vinnikov '06)

Let  $r : G \subset S^d \rightarrow S$  denote a noncommutative rational function on a domain  $G$  containing 0. If  $r$  is matrix convex near 0, then  $r$  has a realization of the form

$$r(X) = r_0 + L(X) + \Lambda(X)^*(1 - \Gamma(X))^{-1}\Lambda(X)$$

for a scalar  $r_0$ , a real linear function  $L$ ,  $\Lambda$  affine linear, and  $\Gamma(X) = \sum A_i \otimes X_i$  for self-adjoint matrices  $A_i$ .

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What structural features appear in the proofs of automatic analyticity theorems?

- Structured domains
- Local boundedness
- Controlled analyticity on 1 dimensional slices - local domination by derivatives, e.g.
- Amenability to approximation - closure under convolution, e.g.

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# A royal proof: dominions

A **dominion** is a set of domains,  $\mathcal{G}$ , in  $\mathbb{R}^d$  satisfying:

- **Translation invariance:** For all  $G \in \mathcal{G}$  and  $r \in \mathbb{R}^d$ ,  $G + r \in \mathcal{G}$ .
- **Scale invariance:** If  $t > 0$  and  $G \in \mathcal{G}$ ,  $tG \in \mathcal{G}$ .
- **Closure under intersection:** For all  $G, H \in \mathcal{G}$ ,  $G \cap H \in \mathcal{G}$ .
- **Locality:** For any  $x \in \mathbb{R}^d$  and  $\varepsilon > 0$ ,  $B_{\mathbb{R}}(x, \varepsilon) \in \mathcal{G}$ .

The class of convex sets in  $\mathbb{R}^d$  is an example of a dominion, as is the class of all open sets in  $\mathbb{R}^d$ .

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# A royal proof: sovereign classes

A **sovereign class**  $\mathcal{F}$  is a set of functions on domains contained in a dominion  $\mathcal{G}$  satisfying:

- **Functions:** For all  $G \in \mathcal{G}$ ,  $\mathcal{F}(G)$  is a set of locally bounded measurable functions.
- **Closure under localization:** If  $f \in \mathcal{F}(G)$  and  $H \subseteq G$  then  $f|_H \in \mathcal{F}(H)$ .
- **Closure under convolution:** The set of functions  $\mathcal{F}(G)$  is convex and closed under pointwise weak limits.
- **Local boundedness:** Each  $f \in \mathcal{F}$  is locally bounded and measurable on finite dimensional affine subspaces on each level.
- **One variable knowledge:** If  $a_i \leq b_i$  for each  $i$ , then  $f_{ab}(t) := f\left(\frac{1-t}{2}a + \frac{1+t}{2}b\right)$  analytically continues to  $\mathbb{D}$  as a function of  $t$ .
- **Control** There is a map  $\gamma$  taking each pair  $(x, f)$  to a non-negative number satisfying
  - 1 For each  $\varepsilon > 0$  there is a universal constant  $c(\varepsilon)$  such that  $\inf_{X \in B_{\mathbb{R}}(x, \varepsilon)} \gamma(x, f) \leq c(\varepsilon) \|f\|_{B_{\mathbb{R}}(x, \varepsilon)}$ .
  - 2 There is a universal positive valued function  $e$  on  $\mathbb{R}^+$  satisfying the following. Write  $f_{ab}(t) = \sum a_n t^n$ . Then,  $\|a_n\| \leq \gamma(x, f)e(\|b - a\|)$ .
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## Theorem (Pascoe-T.D. 2019)

*Every function in a sovereign class is real analytic.*

Essentially forced by choosing the correct axioms.

In Pascoe-T.D. 2020 (*Regal Path*), a simplified set of axioms can be used to extract a proof of the commutative Löwner theorem (Agler-McCarthy-Young 2012 and Pascoe 2018) from the noncommutative theorem (Pascoe-T.D. 2016, Palfia 2020, Pascoe 2019)

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# A road to realizations

Define the **positive-orthant norm of the  $n$ -th derivative at  $X$**  by

$$\|D^n f(X)\|_+ = \sup_{\|H\|=1, H>0, m} \|D^n f(X^{\oplus m})[H]\|.$$

If the  $n$ -th derivative does not exist in some positive direction, we formally set  $\|D^n f(X)\|_+ = \infty$ .

## Theorem

*Matrix monotone functions are a sovereign class.*

Controlled by

$$\gamma(X, f) = \|f(X)\| + \|Df(X)\|_+.$$

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Let  $f$  be a locally bounded matrix convex function defined on some matrix convex set of self adjoints containing 0. There are self-adjoint  $T_i$ , vector  $Q_i$ , a scalar  $a_0$ , and a linear function  $L$  such that

$$f(X) = a_0 + L(X) + \left(\sum Q_i X_i\right)^* (I - \sum T_i X_i)^{-1} \left(\sum Q_i X_i\right).$$

The minimal such realization is essentially unique.



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Thank you!